COMPSCI 323 / ECON 336 / COMPSCI 590.8 Computational Microeconomics

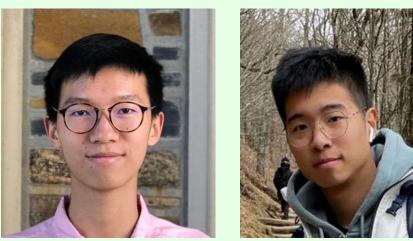
conitzer@cs.duke.edu

https://courses.cs.duke.edu/spring22/compsci323d/

TAs: Chad Kalil, Alex Whitefield; Albert Sun, David (Ruoyu) Wu, Joe (Yuncong) Zuo







CS-ECON

http://econ.cs.duke.edu

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Exploring the Intersection of Computer Science and Economics

Home

Schedule Past Talks

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Reading Group

Who Are We?

We are a group of Duke University faculty, postdocs, and students interested in the intersection of computer science and economics (and the social sciences more broadly) and the impact of this interplay on decisions in information technology and digital business. This includes applying techniques from computer science and optimization to economics -- for example, using computation to design market clearing mechanisms and to implement efficient allocation and pricing in them -- as well as applying techniques from economics to computer science -for example, designing incentives for users of networked computer systems and social networks.

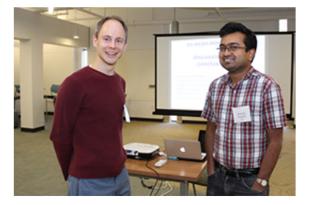
Contacts

For organizational questions about the seminar series:

- Yuan Deng
- <u>Catherine Moon</u>

For other matters, contact the relevant faculty member(s):

- Atila Abdulkadiroglu (Econ)
- Minsent Continue (CC)



CS-Econ Talks

- <u>Upcoming Talks</u>
- <u>Past Talks</u>

Related Seminars

- AI Group (CS)
- <u>Algorithms Seminar</u> (CS)
- <u>Decision Sciences Seminar</u> (Fuqua)
- <u>Duke Robotics, Intelligence,</u> and Vision (DRIV) Seminar

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	2019 Triangle Econometrics Conference									

For Prospective Students

- 🔻 Degree Programs
- M.A. Economics
- M.A. Analytical Political Economy

M.S. Economics & Computation

M.S. Quantitative Financial Economics

For Current Students

• EcoTeach: Student Services Center

M.A. Program Assistant Addie Stagg

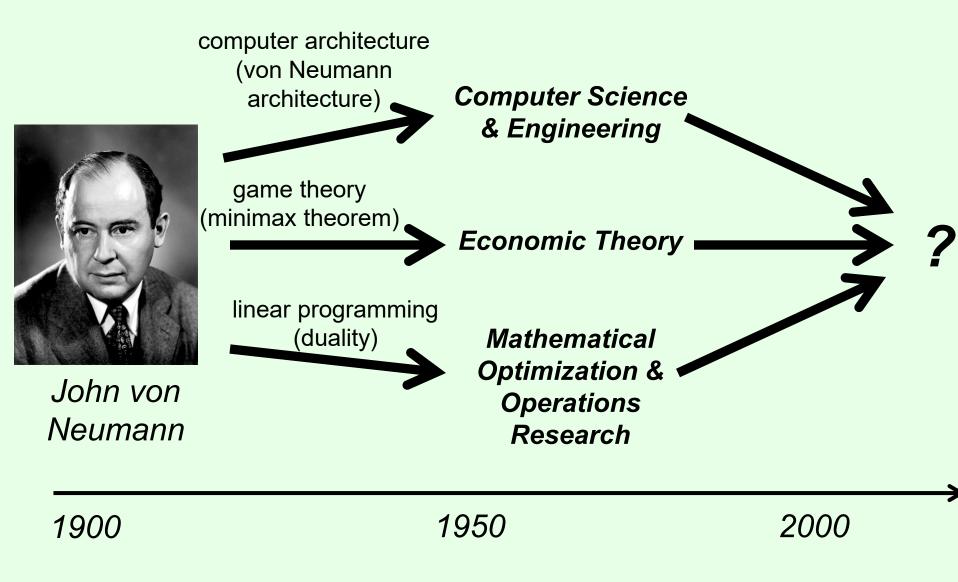
M.S. Economics & Computation

The joint field of economics and computer science has emerged from two converging intellectual needs: Computer science has become increasingly important for economists working with big data to address complex questions. Students interested in learning about computational mechanism design with applications to economics are ideal candidates for this program. Students whose interest is more generally focused on data analytics across a broad range of fields may also be interested in Duke's <u>Master</u> of <u>Quantitative Management</u> (MQM) program, offered at the Fuqua School of Business, and/or Duke's new <u>Master in Interdisciplinary Data Science</u> (MIDS) program, which is accepting its first class in Fall 2018.

The MSEC program combines the strengths of the Departments of Economics and <u>Computer Science</u> to educate students in these important computational skills linked to economics, and to prepare them for Ph.D. studies or careers in economics, finance, government, and business. Reflecting this strong interdisciplinary relationship, Duke University <u>ranks No. 4 for research in economics and computation</u>, according to CSRankings.org.

This program is designed to meet the needs of students with varied levels of exposure to either field, but a strong quantitative background is recommended.

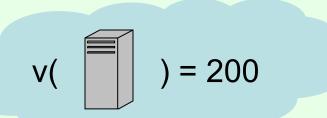
History



What is Economics?

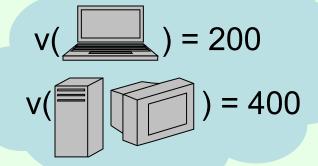
- "a social science that studies the production, distribution, and consumption of goods and services." [Wikipedia, Jan. 2022]
- Some key concepts:
 - Economic agents or players (individuals, households, firms, bots, ...)
 - Agents' current endowments of goods, money, skills, ...
 - Possible outcomes ((re)allocations of resources, tasks, ...)
 - Agents' preferences or utility functions over outcomes
 - Agents' beliefs (over other agents' utility functions, endowments, production possibilities, ...)
 - Agents' possible decisions/actions
 - Mechanism that maps decisions/actions to outcomes

An economic picture



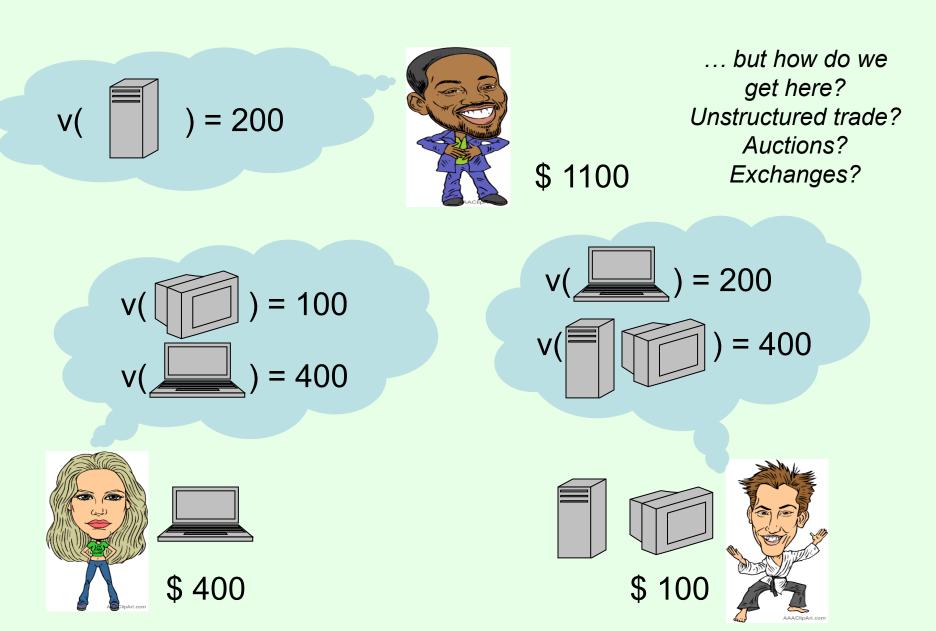


\$ 600





After trade (a more efficient outcome)



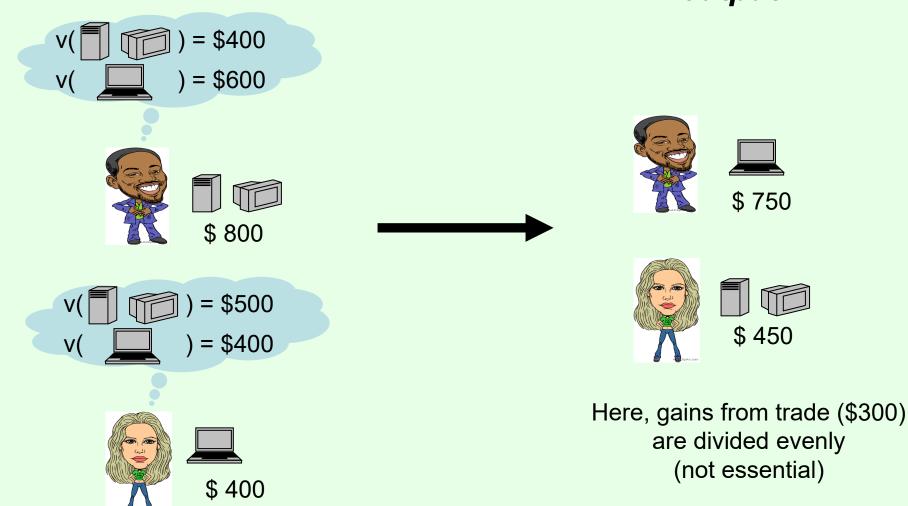
Some distinctions in economics

- Descriptive vs. normative economics
 - Descriptive:
 - seeks only to describe real-world economic phenomena
 - does not care if this is in any sense the "right" outcome
 - Normative:
 - studies how people "should" behave, what the "right" or "best" outcome is
- Microeconomics vs. macroeconomics
 - Microeconomics: analyzes decisions at the level of individual agents
 - deciding which goods to produce/consume, setting prices, ...
 - "bottom-up" approach
 - Macroeconomics: analyzes "the sum" of economic activity
 - interest rates, inflation, growth, unemployment, government spending, taxation, ...
 - "big picture"

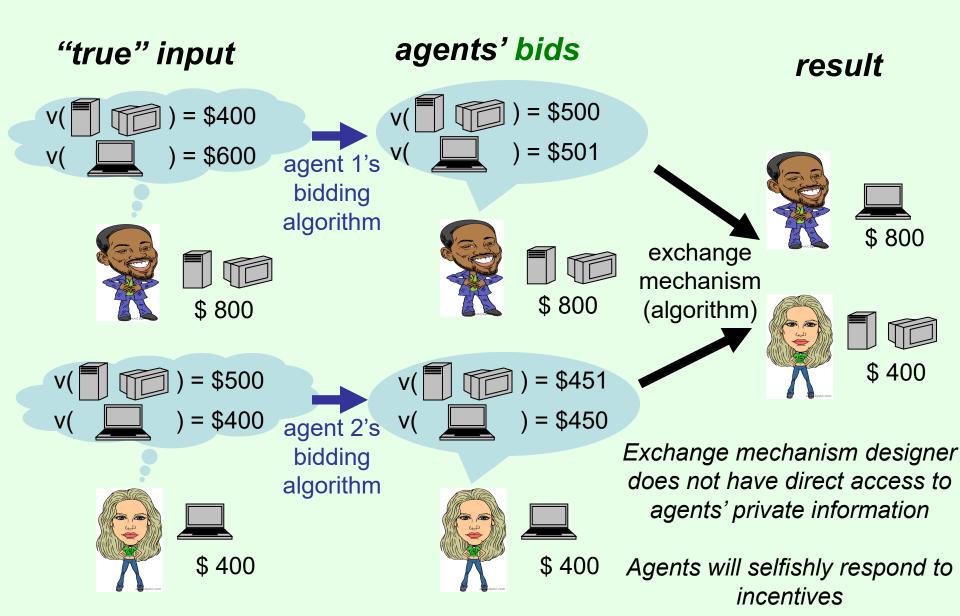
What is Computer Science?

- "Computer science is the study of computation, automation, and information. Computer science spans theoretical disciplines, such as algorithms, theory of computation, and information theory, to practical disciplines including the design and implementation of hardware and software. Computer science is generally considered an area of academic research and distinct from computer programming." [Wikipedia, Jan. 2022]
- A computational problem is given by a function f mapping inputs to outputs
 - For integer x, let f(x) = 0 if x is prime, 1 otherwise
 - For initial allocation of resources + agent utilities x, let f(x) be the (re)allocation that maximizes the sum of utilities
- An algorithm is a fully specified procedure for computing f
 - E.g., sieve of Eratosthenes
 - A correct algorithm always returns the right answer
 - An efficient algorithm returns the answer fast
- Computer science is also concerned with building larger artifacts out of these building blocks (e.g., personal computers, spreadsheets, the Internet, the Web, search engines, artificial intelligence, ...)

Resource allocation as a computational problem (Part 1 of the course) input output



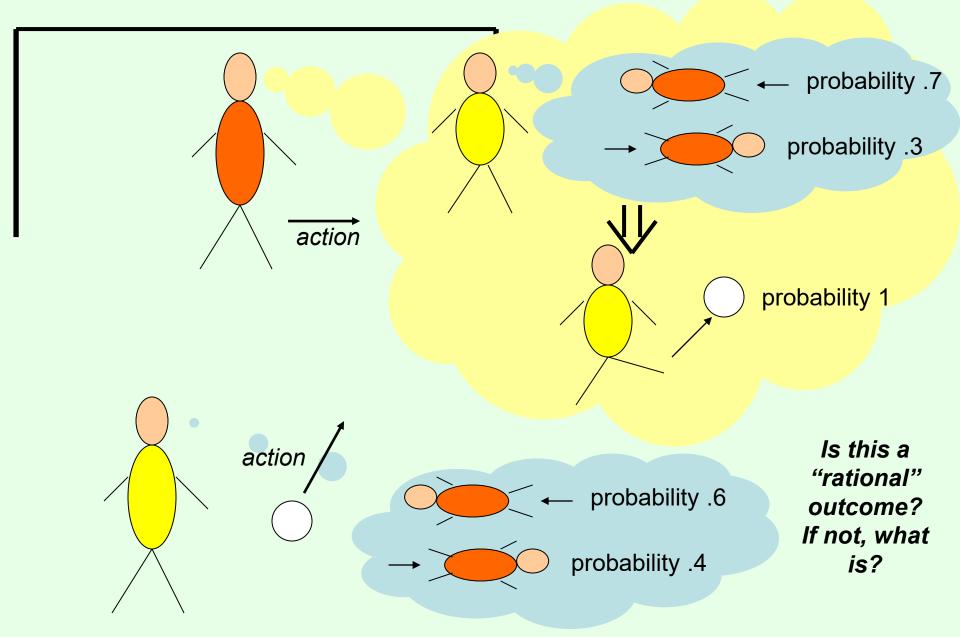
Economic mechanisms



Game theory (Part 2 of the course)

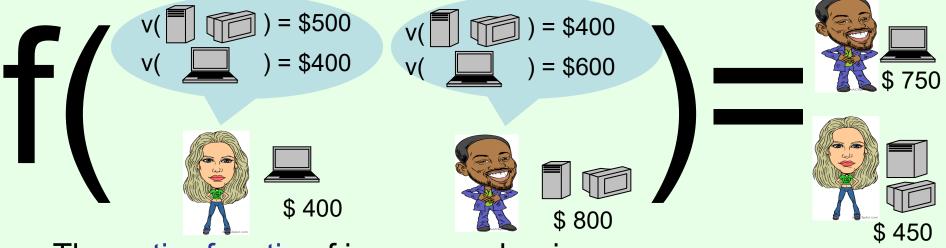
- Game theory studies settings where agents each have
 - different preferences (utility functions),
 - different actions that they can take
- Each agent's utility (potentially) depends on all agents' actions
 - What is optimal for one agent depends on what other agents do
 - Very circular!
- Game theory studies how agents can rationally form beliefs over what other agents will do, and (hence) how agents should act
 - Useful for acting as well as predicting behavior of others

Penalty kick example



Mechanism design (Part 3 of the course)

- Mechanism = rules of auction, exchange, ...
- A function that takes reported preferences (bids) as input, and produces outcome (allocation, payments to be made) as output



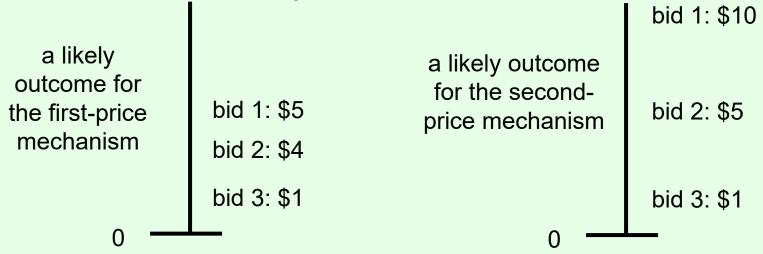
- The entire function f is one mechanism
- E.g., the mechanism from part 1: find allocation that maximizes (reported) utilities, distribute (reported) gains evenly
- Other mechanisms choose different allocations, payments

Example: (single-item) auctions

- Sealed-bid auction: every bidder submits bid in a sealed envelope
- First-price sealed-bid auction: highest bid wins, pays amount of own bid
- Second-price sealed-bid auction: highest bid wins, pays amount of second-highest bid

Which auction generates more revenue?

- Each bid depends on
 - bidder's true valuation for the item (utility = valuation payment),
 - bidder's beliefs over what others will bid (\rightarrow game theory),
 - and... the auction mechanism used
- In a first-price auction, it does not make sense to bid your true valuation
 - Even if you win, your utility will be 0...
- In a second-price auction, (we will see later that) it always makes sense to bid your true valuation



Are there other auctions that perform better? How do we know when we have found the best one?

Mechanism design...

- Mechanism = game
- → we can use game theory to predict what will happen under a mechanism

- if agents act strategically

- When is a mechanism "good"?
 - Should it result in outcomes that are good for the reported preferences, or for the true preferences?
 - Should agents ever end up lying about their preferences (in the game-theoretic solution)?
 - Should it always generate the best allocation?
 Should agents ever burn money?(!?)
- Can we solve for the optimal mechanism?

How are we going to solve these problems? (Part 0)

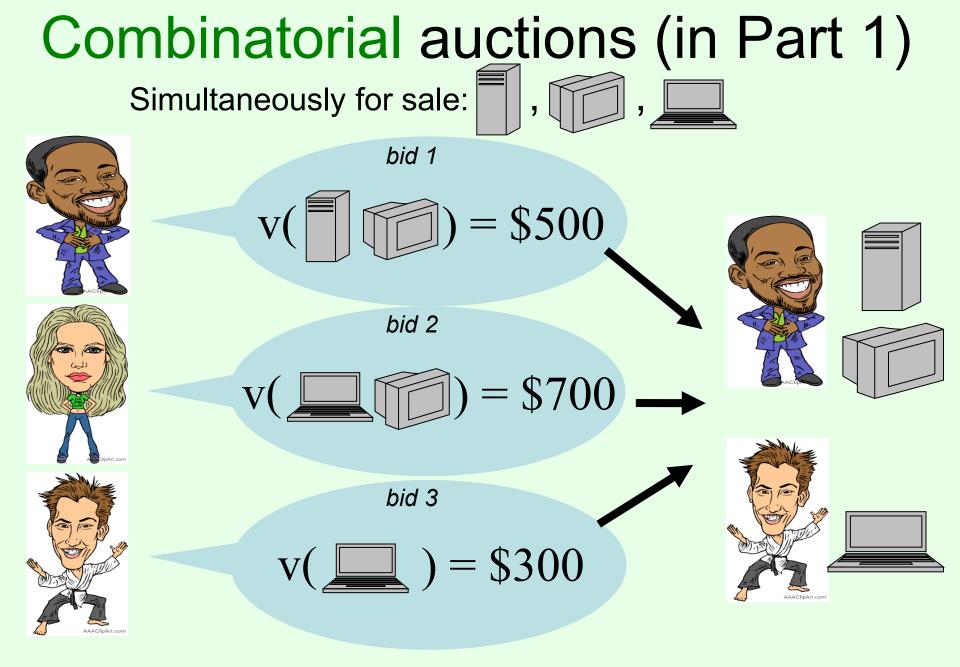
• This is not a programming course

- Will use optimization software
 - GNU Linear Programming Kit (GLPK)
 - Linear programming, mixed integer linear programming

Uses of LP, MIP in this course

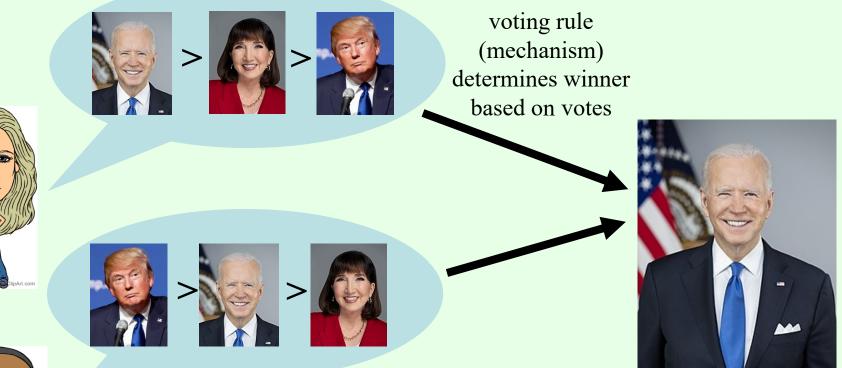
	Linear programming	Mixed integer linear programming
Part 1 (expressive marketplaces)	Winner determination in auctions, exchanges, with partially acceptable bids	Winner determination in auctions, exchanges, without partially acceptable bids
Part 2 (game theory)	Dominated strategies Minimax strategies Correlated equilibrium Optimal mixed strategies to commit to	Nash equilibrium
Part 3 (mechanism design)	Automatically designing optimal mechanisms that use randomization	Automatically designing optimal mechanisms that do not use randomization

Other settings/applications



used in truckload transportation, industrial procurement, radio spectrum allocation, ...

Voting (in Part 1)

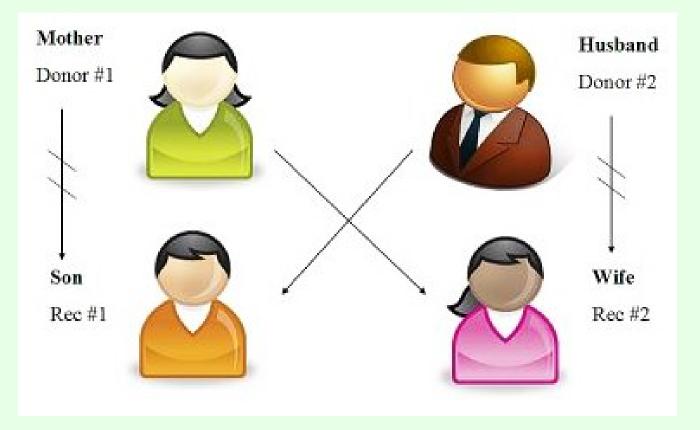




- Can vote over other things too
 - Where to go for dinner tonight, other joint plans, ...
- Many different rules exist for selecting the winner

Kidney exchange (in Part 1)

• Kidney exchanges allow patients with willing but incompatible live donors to swap donors



Kidney exchange (in Part 1)

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Prescription Al

This series explores the promise of AI to personalize, democratize, and advance medicine—and the dangers of letting machines make decisions.

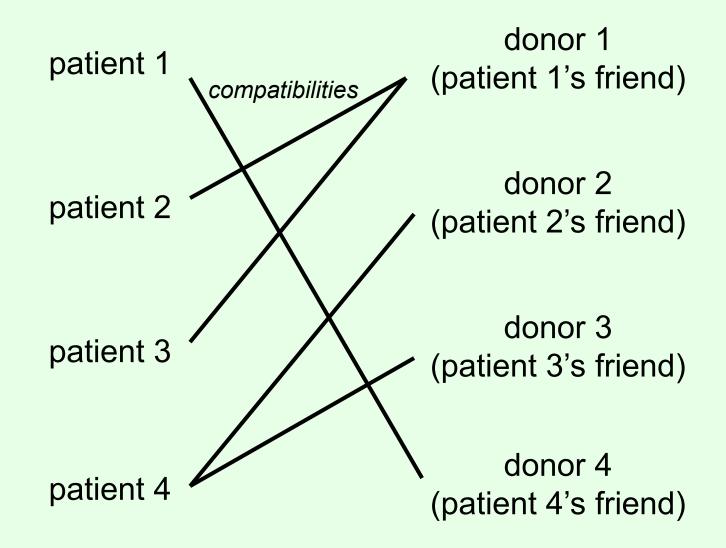
THE BOTPERATING TABLE

How AI changed organ donation in the US

By Corinne Purtill • September 10, 2018

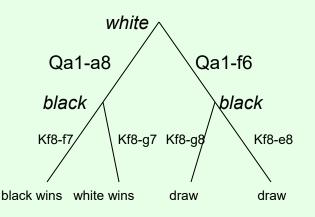
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Kidney exchange (in Part 1)

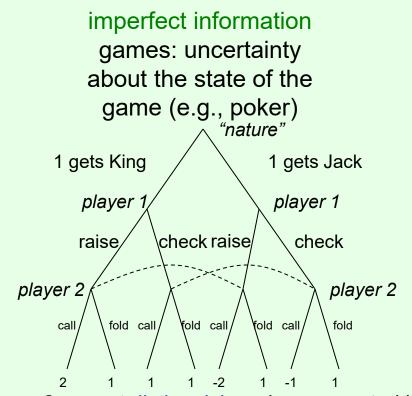


Game playing & AI (in Part 2)

perfect information games: no uncertainty about the state of the game (e.g. tictac-toe, chess, Go)



- Optimal play: value of each node = value of optimal child for current player (backward induction, minimax)
- For chess and Go, tree is too large
 - Use other techniques (heuristics, limited-depth search, alpha-beta, deep learning, ...)
- Top computer programs better than humans in chess, not yet in Go



- Player 2 cannot distinguish nodes connected by dotted lines
 - Backward induction fails; need more sophisticated game-theoretic techniques for optimal play
- Small poker variants can be solved optimally
- Humans still better than top computer programs at full-scale poker (at least most versions)
- Top computer (heads up) poker players are based on techniques for game theory

Science

2019 BREAKTHROUGH OF THE YEAR

Darkness made visible

RUNNERS-UP

Artificial intelligence masters multiplayer poker

Editorial

fibrosis

Podcast

BREAKDOWNS

Measles resurgent

Bird counts dwindling

An eleventh hour climate awakening?

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Artificial intelligence masters multiplayer poker



JASON SOLO/THE JACKY WINTER GROUP

This year, an artificial intelligence (AI) program beat some of the world's best players in the most popular version of poker, no-limit Texas Hold 'em. The landmark result marks the first time AI has prevailed in a multiplayer contest in which players have only imperfect information about the state of the game.

Al has been trouncing humans in games at a spectacular rate. In 2007, computer scientists developed a program guaranteed not to lose at checkers. In 2016, another team developed an Al program that defeated the best humans at Go, a board game with vastly more configurations than checkers.

Poker presents a stiffer challenge, as players cannot see their opponents' cards and thus have limited information. In 2017, computer scientists developed an AI program unbeatable at a two-player version of Hold 'em—in which each player forms a hand from five cards laid face up on the table and two more each holds privately.

Now, AI has bested world-class players in the full multiplayer game, as computer scientists at Carnegie Mellon University in Pittsburgh, Pennsylvania, announced in August. By playing 1 trillion games against itself, their program, Pluribus, developed a basic strategy for various kinds of situations—say, playing for an inside straight. For each specific hand, it could also think through how the cards would likely play out. In 20,000 hands with six players it outperformed 15 top-level players, as measured by average winnings per hand.



Real-world security applications (in Part 2)





Airport security

Milind Tambe's TEAMCORE group $(USC \rightarrow Harvard)$

Where should checkpoints, canine units, etc. be deployed?

Federal Air Marshals

Which flights get a FAM?





US Coast Guard

Which patrol routes should be followed?

Wildlife Protection

Where to patrol to catch poachers or find

their snares?



Global Presence of Security Games Efforts



Prediction markets

(Jan. 4, 2022)

Predict It	Markets	Insights	Leaderboards	Support	Login	Sign Up 👂
Biden Administration		U.S. Elec	otions	Congress	State/Local	World

Who will win the 2024 US presidential election?

Contract	Latest Yes Price	Best Offer	Best Offer
Donald Trump	30¢ ₁¢♥	31¢ Buy Yes	Buy No 70¢
Joe Biden	22¢ NC	23¢ Buy Yes	Buy No 78¢
Ron DeSantis	20¢ ₁¢♠	20¢ Buy Yes	Buy No 81¢
Kamala Harris	11¢ NC	12¢ Buy Yes	Buy No 89¢
Pete Buttigieg	6¢ 1¢♥	7¢ Buy Yes	Buy No 94¢
The	e Rules	Related Markets	

Prediction markets

(Jan. 4, 2022)

Predict It	Markets	Insights	Leaderboards	Support	L	ogin Sign Up
Biden Administration		U.S. Elections		Congress	State/Local	World

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Predictlt is intended and offered as an experimental research and educational facility of Victoria University of Wellington, New Zealand ("Provider" or "We"), not as an investment market or a gambling facility. PredictIt is not regulated by, nor are its operators registered with, the U.S. Commodity Futures Trading Commission (CFTC) or any other regulatory authority.

Provider has received a no-action-letter from the Division of Market Oversight of the Commodity Futures Trading Commission. Without explicitly asserting jurisdiction over Provider or any of its submarkets, this letter, dated October 29, 2014, extended no-action relief to Provider's Political and Economic Indicator Markets (the latter limited to students, faculty and staff of participating universities). The letters are available at the CFTC website as part of their Freedom of Information Act documents. Pursuant to this letter, there is "a limit of 5000 total traders in any particular contract", and "a limit on investment by any single participant in any particular contract [of] \$850".

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ii. No political party or other organization referred to on the Website in connection with any Market has (or has had) any role in the promotion or operation of this Website or any Market.

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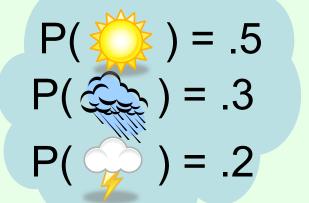
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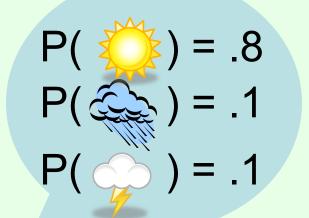
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Financial securities (in Part 1)

- Tomorrow there must be one of
- Agent 1 offers \$5 for a security that pays off
 \$10 if offers or offers
- Agent 2 offers \$8 for a security that pays off \$10 if
- Agent 3 offers \$6 for a security that pays off
 \$10 if
- Can we accept some of these at offers at no risk?

How to incentivize a weather forecaster (in Part 3)

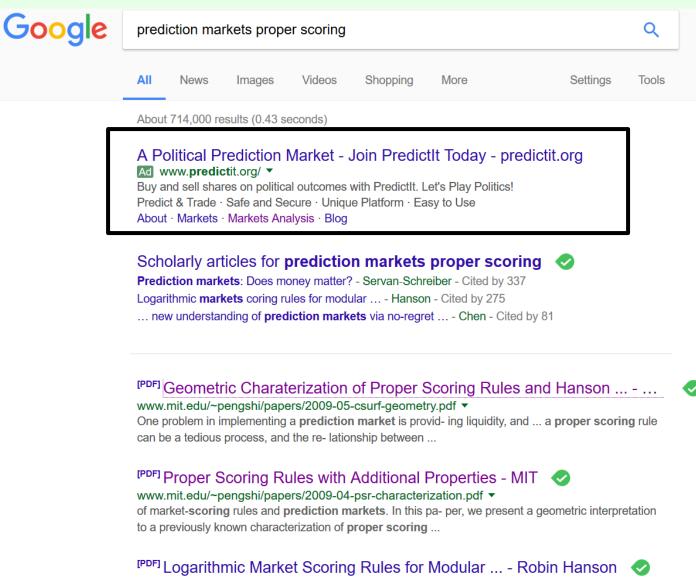






- Forecaster's bonus can depend on
 - Prediction
 - Actual weather on predicted day
- Reporting true beliefs should maximize expected bonus

Sponsored search / ad auctions (in Part 3)



hanson.gmu.edu/mkt**score**.pdf **▼**

by R Hanson - 2002 - Cited by 275 - Related articles

cannot even predict the direction in which others will disagree with them (Hanson, For a non-proper

Choice of ads (if any) to show determined by:

•

- Advertiser
 bid
- Predicted
 likelihood
 of click